

Fast Inverse Square Root (0x5F37591)

$$\text{finding } x_0 + f_x = \underbrace{\frac{1}{2} E_x}_{\text{1 bit}} \underbrace{\pi}_{\text{b bits}} - \underbrace{\frac{M_x}{2^{n-b-1} b \text{ bits}}}_{\text{n-b-1 bits}}$$

Input only defined for $x > 0 \rightarrow \mathbb{R}^+$

$$\therefore f_x = \left(1 + \frac{M_x}{2^{n-b-1}}\right) 2^{E_x - (2^{b-1} - 1)}$$

For simplicity, let $L = 2^{N_b - 1}$, which normalizes the mantissa M to $0 \leq M < 1$
 + let $B = 2^{b-1} - 1$, which is the exponent bias.

$$f_x = \left(1 + \frac{M_x}{L}\right) 2^{E_x - B} \quad (1)$$

$$\text{looking for } y = \frac{1}{\sqrt{x}} = x^{-\frac{1}{2}}$$

Let $f_x + f_y$ be floating-point representation of $x + y$ numerically.

$$\therefore f_y = f_x^{-\frac{1}{2}} \quad \text{ignoring errors introduced by floating}$$

$$\log_2 f_y = \log_2(f_x^{-\frac{1}{2}})$$

$$\log_2 f_x = -\frac{1}{2} \log_2 f_x$$

$$\log_2 \left(\left(1 + \frac{M_x}{L}\right) 2^{E_x - B} \right) = -\frac{1}{2} \log_2 \left(\left(1 + \frac{M_x}{L}\right) 2^{E_x - B} \right)$$

$$\log_2 \left(1 + \frac{M_x}{L} \right) + \log_2 \left(2^{E_x - B} \right) = -\frac{1}{2} \left[\log_2 \left(1 + \frac{M_x}{L} \right) + \log_2 \left(2^{E_x - B} \right) \right]$$

$$\left(1 + \frac{M_x}{L}\right) + E_x - B = -\frac{1}{2} \log_2 \left(1 + \frac{M_x}{L} \right) - \frac{1}{2} E_x + \frac{1}{2} B$$

$$\log_a(1+\frac{L}{2}) + E_y = \alpha \theta_x + \theta_y - \frac{1}{2}E_x + \theta_z$$

$$\log_a(1+\frac{L}{2}) + E_y = \alpha \log_a(1+\frac{L}{2}) - \frac{1}{2}(E_x - \theta_z) \quad (2)$$

Now we can bring eq. on $\log_a(1+L/2) = 1$:



$$\log_a(1+L/2) = 1$$

or $\log_a(1+L/2) = 0.693$

then form $L = 25m$

Substituting (3) into (2) we have:

$$\frac{M_y}{L} + \theta_y = E_y = -\frac{1}{2}\left(\frac{M_y}{L} + \theta_x\right) - \frac{1}{2}(E_x - \theta_z)$$

$$\frac{M_y}{L} + E_y = -\frac{1}{2}\theta_x - \frac{1}{2}\theta_z - \frac{1}{2}E_x$$

$$E_y L + M_y = -\frac{1}{2}L(\frac{1}{2}\theta_x + \theta_y) - \frac{1}{2}E_x L$$

$$- \frac{1}{2}M_x - L(\frac{1}{2}\theta_x + \theta_y) - \frac{1}{2}E_x L$$

$$E_y L + M_y = \frac{1}{2}L - (\frac{1}{2}\theta_x + \theta_y) - \frac{1}{2}(E_x L + M_x)$$

Plane 1: the right look at θ_x and E_x

$N=1$

$$2^{N-1} = \Sigma L - M_x$$

x is we take I_x and T_x to be the 1st
 $f_x + f_y$, respectively, (not the right), then we have

$$I_x = E_x L + M_x \quad , \quad I_y = E_y L + M_y.$$

So from (8), (ii) gives us

$$I_y = L \left(\frac{3}{2} B - \left(\frac{1}{2} \theta_x + \theta_y \right) \right) - \frac{1}{2} I_x$$

or more simply

$$I_y = R - \frac{1}{2} I_x$$

$$\text{where } R = L \left(\frac{3}{2} B - \left(\frac{1}{2} \theta_x + \theta_y \right) \right)$$

And that's the basic technique: take floating point
 integer division (of $\theta_x + \theta_y$ by 2), & subtract
 our magic integer R . If you do it you take $\theta_x + \theta_y$
 as a floating point just get R .

The error comes from the fact that $\theta_x + \theta_y$ is not
 a specific values of x & y (it's not an integer).

So we need to use $f_x + f_y$ as best as we can
 be an approximation for most x & y .

For example, if we take

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$$= 2^{23} \left(\frac{3}{2} (R) - \left(\frac{1}{2} \theta_x + \theta_y \right) \right)$$